# Molecular Dynamics

Goal: trajectories x(t) that sample the different conformations of the system so that we can calculate correct **thermodynamic observables**.

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# Molecular Dynamics Algorithm

equations of motion for N particles

$$\mathbf{F}_i = m_i \mathbf{a}_i = m_i \frac{d^2 \mathbf{r}}{dt^2} = m_i \ddot{\mathbf{r}}_i$$

potential energy

$$U(\mathbf{r}_1, \dots, \mathbf{r}_N) = U_{\text{bonded}}(\{\mathbf{r}_i\}) + U_{\text{nonbonded}}(\{\mathbf{r}_i\})$$

$$U_{\text{bond}}(\mathbf{r}_j, \mathbf{r}_k) = \frac{K}{2} (|\mathbf{r}_j - \mathbf{r}_k| - b_0)^2$$

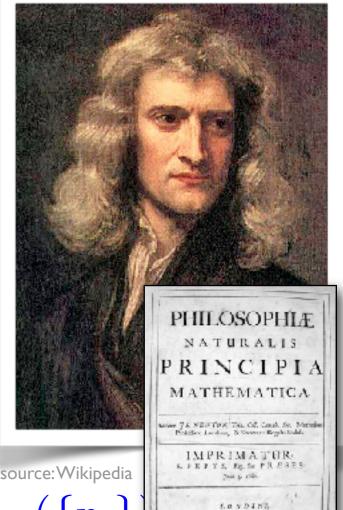
$$U_{\text{angles}}(\mathbf{r}_i, \mathbf{r}_j, \mathbf{r}_k) \quad U_{\text{dih}}(\mathbf{r}_i, \mathbf{r}_j, \mathbf{r}_k, \mathbf{r}_l)$$

$$U_{\text{elec}}(\mathbf{r}_{j}, \mathbf{r}_{k}) = \frac{1}{4\pi\epsilon_{0}} \frac{q_{j}q_{k}}{|\mathbf{r}_{j} - \mathbf{r}_{k}|}$$

$$U_{\text{vdW}}(\mathbf{r}_{j}, \mathbf{r}_{k}) = 4\varepsilon \left[ \left( \frac{\sigma}{r_{jk}} \right)^{12} - \left( \frac{\sigma}{r_{jk}} \right)^{6} \right]$$

forces

$$\mathbf{F}_i = -\nabla_i U(\mathbf{r}_1, \dots, \mathbf{r}_i, \dots, \mathbf{r}_N)$$



#### Basic MD program

```
initialize()
t=0
while t < t_{\text{max}}
        \mathbf{f} = forces(\mathbf{r})
        \mathbf{r} = integrate(\mathbf{r}, \mathbf{f})
        write_trajectory(r)
        t = t + dt
```

#### **Basic workflow**

- I. prepare starting conformation
- 2. run equilibration
- 3. run production
  - compute estimators for observables  $a(\mathbf{r}, \mathbf{v})$
  - save trajectory (r, maybe v and F)
- 4. analyse trajectory/output files

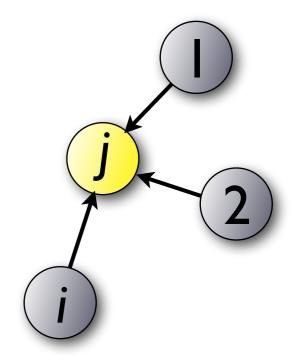
#### Force calculation

```
initialize()
t=0
while t < t_{\text{max}}
       f = forces(r)
       \mathbf{r} = integrate(\mathbf{r}, \mathbf{f})
       write_trajectory(r)
       t = t + dt
```

#### Force calculation

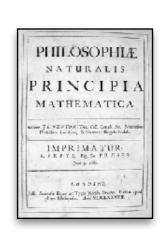
$$\mathbf{F}_i = -\nabla_i U(\mathbf{r}_1, \dots, \mathbf{r}_i, \dots, \mathbf{r}_N)$$

For pair-potentials  $v(|\mathbf{r}_j - \mathbf{r}_i|) = v(r_{ij})$ 



$$\mathbf{F}_{j} = \sum_{i} \left[ -\nabla_{i} v(r_{ij}) \right] = -\sum_{i} \frac{\mathbf{r}_{ij}}{r_{ij}} \frac{\partial}{\partial r_{ij}} v(r_{ij})$$

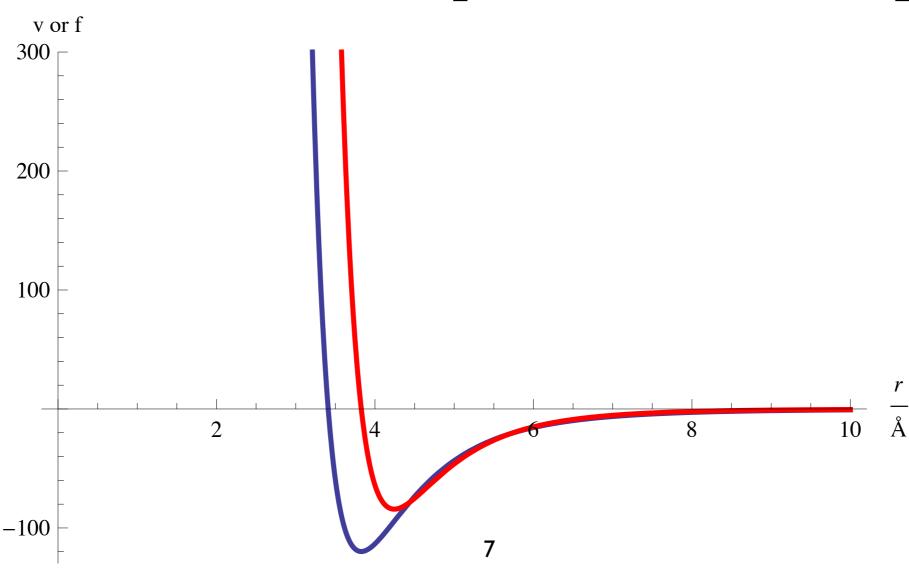
$$\mathbf{f}_{ij} = -\mathbf{f}_{ji}$$
 (Newton's 3rd law)



#### Lennard-Jones Potential

$$v(r_{ij}) = 4\epsilon \left[ \left( \frac{\sigma}{r_{ij}} \right)^{12} - \left( \frac{\sigma}{r_{ij}} \right)^{6} \right]$$

$$\mathbf{f}(r_{ij}) = -24\epsilon \frac{\mathbf{r}_{ij}}{r_{ij}^2} \left[ 2\left(\frac{\sigma}{r_{ij}}\right)^{12} - \left(\frac{\sigma}{r_{ij}}\right)^{6} \right]$$



## Integrating the Equations of Motion

```
initialize()
t=0
while t < t_{\text{max}}
        \mathbf{f} = forces(\mathbf{r})
        \mathbf{r} = integrate(\mathbf{r}, \mathbf{f})
        write_trajectory(r)
       t = t + dt
```

# Integrating the Equations of Motion

$$m_i \ddot{\mathbf{r}}_i = \mathbf{F}_i$$

- have forces  $\mathbf{F}_i$  and current positions  $\mathbf{r}_i(t)$
- want: positions  $\mathbf{r}_i(t+\Delta t)$

## Velocity Verlet Integrator

$$\mathbf{v}(t + \frac{\Delta t}{2}) = \mathbf{v}(t) + \frac{\Delta t}{2} \frac{\mathbf{F}(t)}{m} \tag{1}$$

$$\mathbf{r}(t + \Delta t) = \mathbf{r}(t) + \Delta t \,\mathbf{v}(t + \frac{\Delta t}{2}) \tag{2}$$

$$\mathbf{v}(t + \Delta t) = \mathbf{v}(t + \frac{\Delta t}{2}) + \frac{\Delta t}{2} \frac{\mathbf{F}(t + \Delta t)}{m}$$
(3)

- gives both positions and velocities
- one force evaluation per time step
- see Module 10 ODEs

#### Good integrators

- speed? not very relevant
- accuracy for large time steps  $\Delta t$
- energy conservation (for  $\partial H/\partial t = 0$ )
  - short term
  - long term (more important)
- time reversible (Newton's EOMs!)
- phase-space area preserving (Hamiltonian dynamics!)
- accurately predict the trajectories of all particles for short and long times?

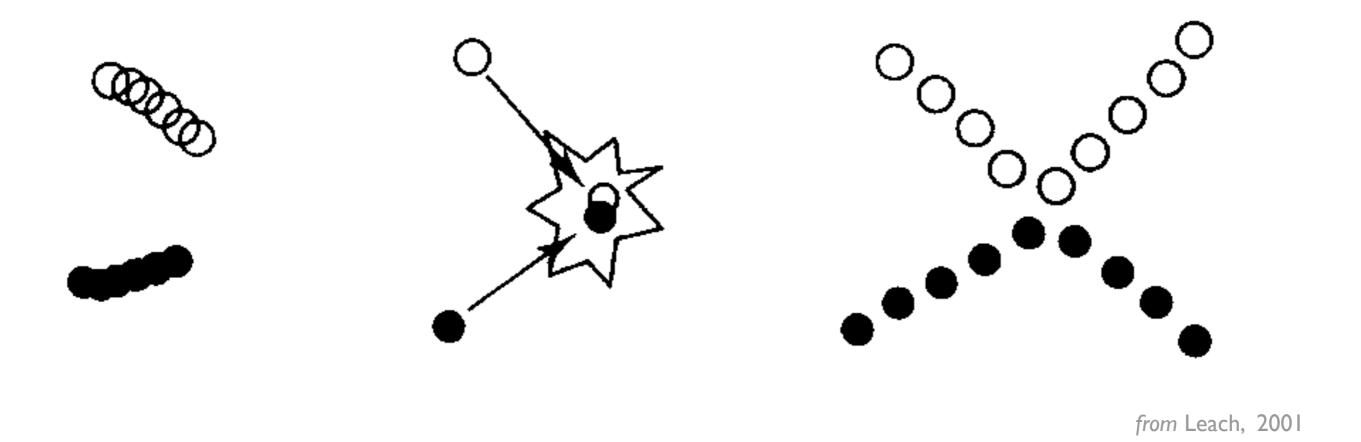
#### Good integrators

- speed/memory? not very relevant
- accuracy for large time steps  $\Delta t$
- energy conservation (for  $\partial H/\partial t = 0$ )
  - short term
  - long term (more important)
- time reversible (Newton's EOMs!)
- phase-space area preserving (Hamiltonian dynamics!)
- accurately predict the trajectories of all particles for short and long times?

#### Verlet

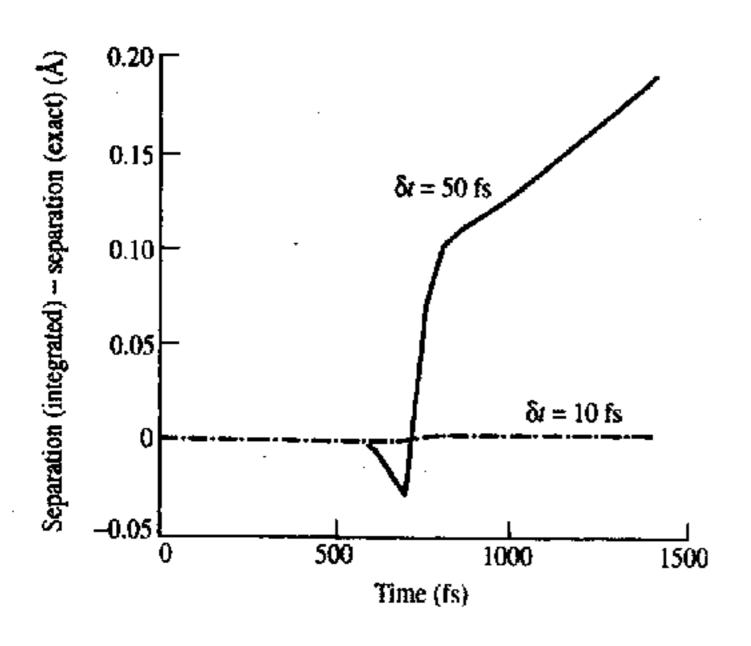
- fast, small memory requirements
- not very high accuracy for long time steps
- fair short- and good long term energy conservation
- time reversible and area preserving ("symplectic")

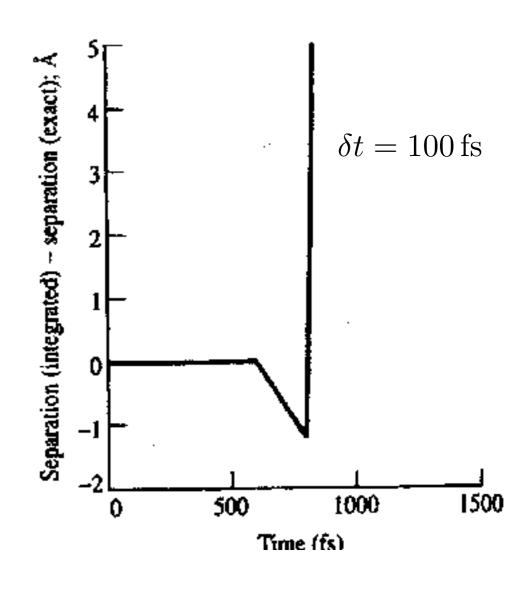
#### Time step

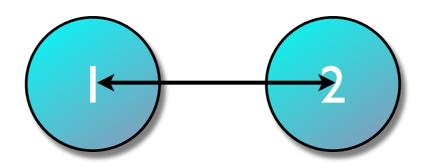


 rule of thumb for Verlet-type integrators: 5 steps per period (typically I-2 fs in biomolecular systems)

#### Time step







## Thermodynamic observables

• Macroscopic thermodynamic quantities (temperature *T*, pressure *P*, volume *V*, particle number *N*, heat capacities, dipole moments, magnetization, ...) can be written as **averages** over functions that depend on microscopic positions and velocities ("estimator").

$$A = \langle a \rangle$$

$$A = \langle a(\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_N, \mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_N) \rangle$$

$$\mathbf{A} = \langle a(\mathbf{x}_1(t), \mathbf{x}_2(t), \dots, \mathbf{x}_N(t), \mathbf{v}_1(t), \mathbf{v}_2(t), \dots, \mathbf{v}_N(t)) \rangle_t$$

$$\mathbf{A} = \frac{1}{N_t} \sum_{i=1}^{N_t} a(\mathbf{x}_1(t), \mathbf{x}_2(t), \dots, \mathbf{x}_N(t), \mathbf{v}_1(t), \mathbf{v}_2(t), \dots, \mathbf{v}_N(t))$$

## **Energy** E

$$E = \langle \mathcal{H} \rangle$$

$$\mathcal{H} = T_{\text{kin}}(t) + U(t)$$

$$= \sum_{i=1}^{N} \frac{1}{2} m_i \mathbf{v}_i^2 + \sum_{i < j}^{N-1} U_{\text{LJ}}(r_{ij})$$

$$r_{ij} := |\mathbf{r}_j - \mathbf{r}_i|$$

H: "Hamiltonian" (total instantaneous energy)

Energy conservation: 
$$\frac{d\mathcal{H}}{dt} = 0$$

## Temperature T

$$T_{\text{kin}}(t) = \frac{1}{2} \sum_{i=1}^{N} m_i \, \mathbf{v}_i(t)^2 \qquad \langle T_{\text{kin}} \rangle = N \frac{3}{2} k_B T$$

$$\langle T_{\rm kin} \rangle = N \frac{3}{2} k_B T$$

1/2 kT per degree of freedom

$$T=rac{2\langle T_{
m kin}
angle}{3k_BN}$$

#### Boltzmann's constant

$$k_B = 1.38064852 \times 10^{-23} \,\mathrm{J \cdot K^{-1}}$$

$$\mathcal{T}(t) = rac{2T_{\mathrm{kin}}(t)}{3k_B N}$$
  $T = \langle \mathcal{T}(t) 
angle_t$ 

$$\mathbf{A} = \frac{1}{N_t} \sum_{i=1}^{N_t} a(\mathbf{x}_1(t), \mathbf{x}_2(t), \dots, \mathbf{x}_N(t), \mathbf{v}_1(t), \mathbf{v}_2(t), \dots, \mathbf{v}_N(t))$$

#### Pressure P

 pressure can be derived from a version of the virial theorem (... another time)

"virial" 
$$w:=\left\langle \sum_{i< j}^{N-1}\mathbf{r}_{ij}\cdot\mathbf{f}_{ij}
ight
angle \mathbf{F}_i=\sum_{j\neq i}\mathbf{f}_{ij}$$

$$\mathcal{P}(t) = V^{-1} \left[ Nk_B \mathcal{T}(t) + \frac{1}{3} \sum_{i < j}^{N-1} \mathbf{r}_{ij}(t) \cdot \mathbf{f}_{ij}(t) \right]$$
$$P = \langle \mathcal{P}(t) \rangle_t$$

#### Summary

- outline of the MD algorithm
- use of periodic boundary conditions
- potential truncation
- integrators
- calculating macroscopic observables from microscopic estimators

# **Appendix**

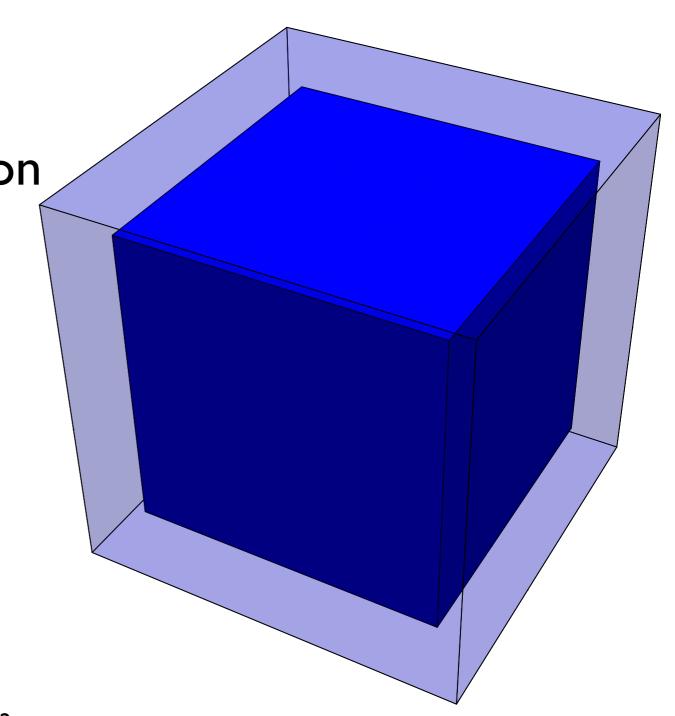
## Small systems: Surface effects

• In a small system, "most" particles are "near: the surface.

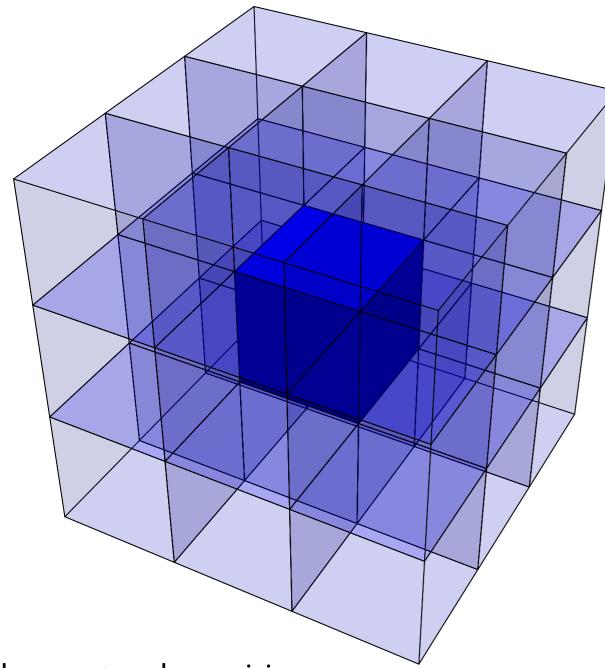
• "near" = typical interaction distance  $d\sim3\sigma$ 

•  $N_{\text{surf}}/N \sim N^{-1/3}$ 

 Problem when we are interested in bulk properties.



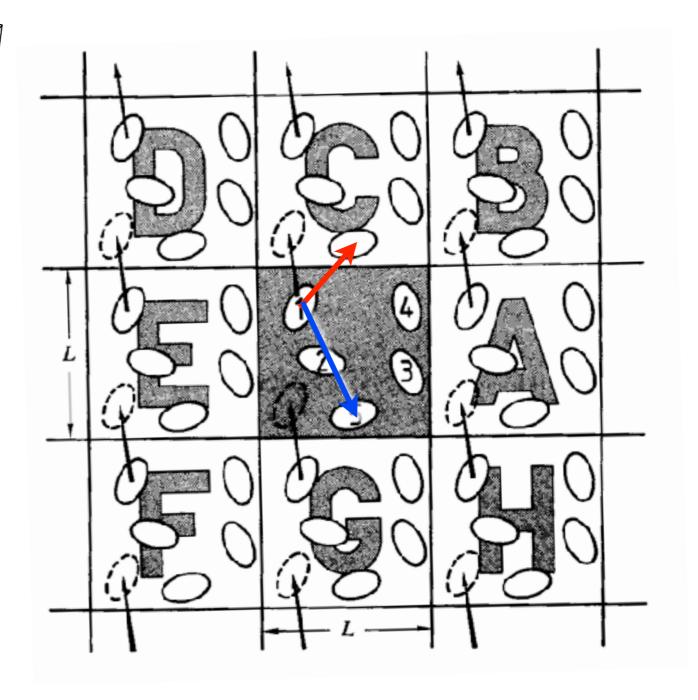
#### Small systems: Periodic boundaries



box centered on origin:

$$\mathbf{r}_i \leftarrow \mathbf{r}_i - L \operatorname{rint}\left(\frac{\mathbf{r}_i}{L}\right)$$

$$\mathbf{r}_{ij} \leftarrow \mathbf{r}_{ij} - L \operatorname{rint}\left(\frac{\mathbf{r}_{ij}}{L}\right)$$



from Allen & Tildesley, 1987

#### Periodic boundaries: Potential Issues

- simulate *infinite* system so must handle an infinite amount of interactions (though often truncation is permissible)
- spurious correlations/ordering
- only fluctuations allowed with lattice periodicity, and max wavelength  $\lambda = L$ 
  - phase transitions with long wavelength fluctuations problematic
- non-isotropic pair distribution function

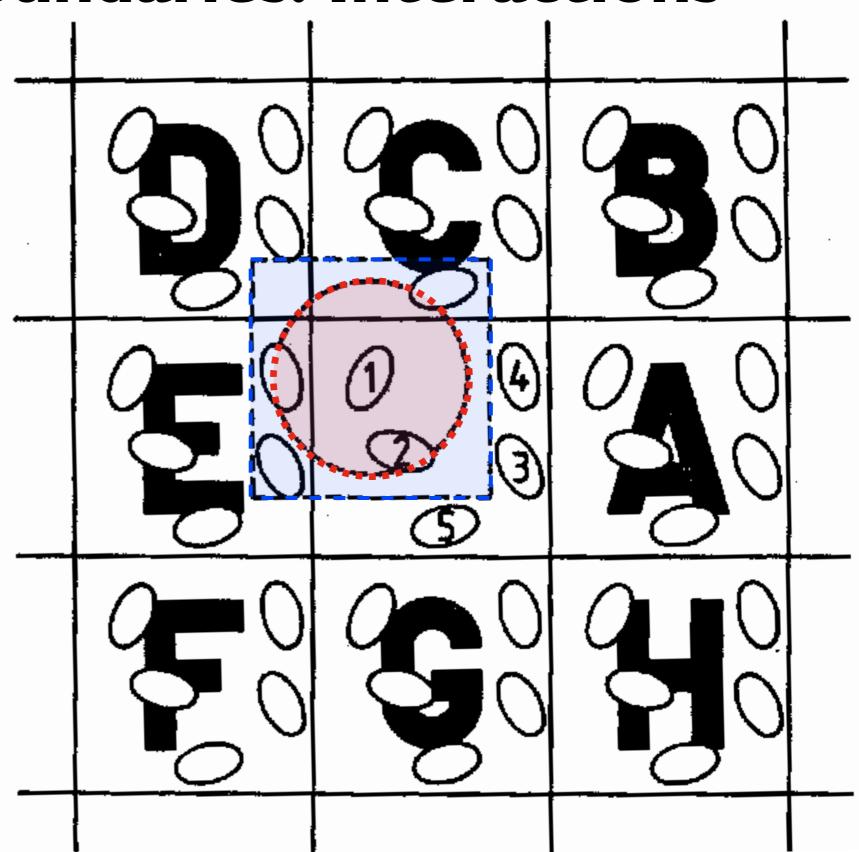
... but generally works rather well!

#### Periodic boundaries: Interactions

- pair wise  $v(r_{ij}) \sim N^2$
- minimum image convention
- truncation at cutoff  $R_c$

v(r) must fall off faster than  $r^{-3}$ 

 $R_c < L/2$ 



#### Truncation of potentials

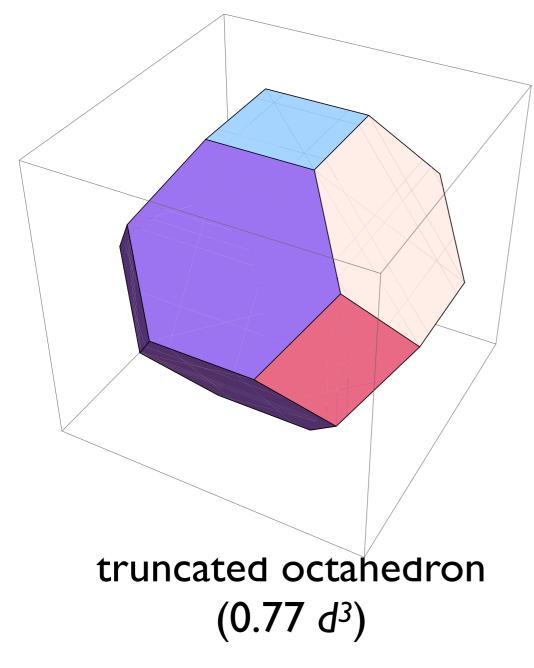
(proper periodic treatment: advanced topic)

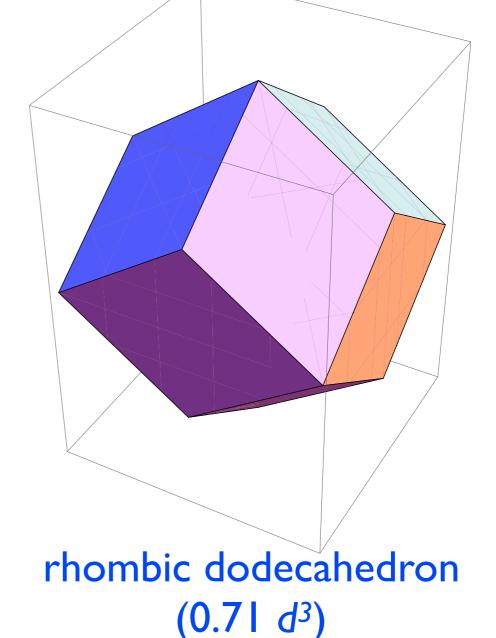
• simple truncation 
$$u^{\mathrm{trunc}}(r) = \left\{ \begin{array}{ll} u(r) & r \leq R_c \\ 0 & r > R_c \end{array} \right.$$

• truncation and shift 
$$u^{\mathrm{shift}}(r) = \left\{ \begin{array}{ll} u(r) - u(R_c) & r \leq R_c \\ 0 & r > R_c \end{array} \right.$$

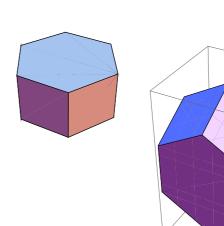
minimum image convention

#### Periodic boundaries: Cells





- + cube/parallel epiped  $(d^3)$
- + hexagonal prism
- + elongated rhombic dodecahedron



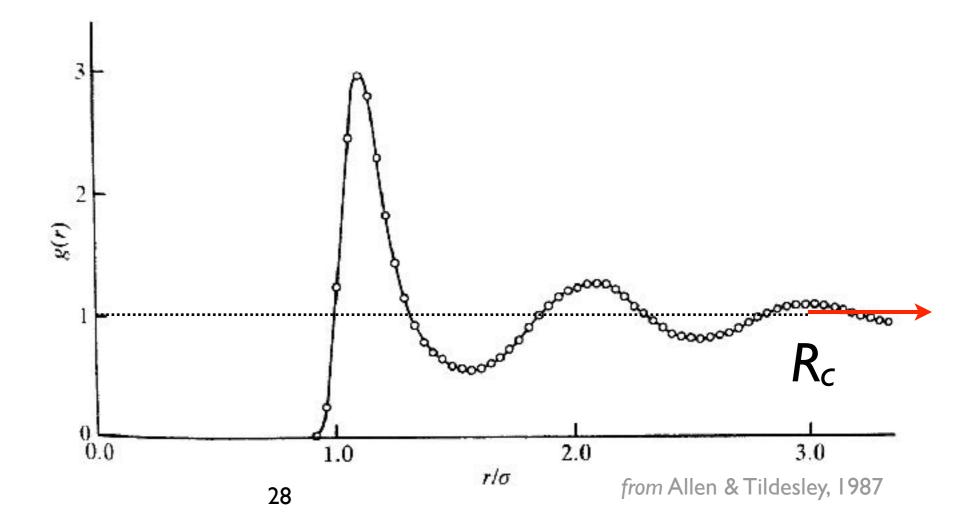
#### Periodic boundaries: Tail correction

$$U_{\text{tot}} = \frac{1}{2} N \rho \int_0^\infty u(r) g(r) 4\pi r^2 dr$$

$$U_{\text{tot}} = \sum_{i < j} u_c(r_{ij}) + \frac{1}{2} N \rho \int_{R_c}^{\infty} u(r) 4\pi r^2 dr$$

$$u_c(r) = \begin{cases} u(r) & r \le R_c \\ 0 & r > R_c \end{cases}$$

$$g(r) \approx 1$$
 for  $r > R_c$ 



## Verlet Integrator

$$\mathbf{r}(t + \Delta t) \approx 2\mathbf{r}(t) - \mathbf{r}(t - \Delta t) + \frac{\mathbf{F}(t)}{m} \Delta t^2$$

- no velocities needed
- same accuracy as velocity Verlet
- velocities can be computed (but  $\mathcal{O}(\Delta t^2)$ )

$$\mathbf{v}(t) = \frac{\mathbf{r}(t + \Delta t) - \mathbf{r}(t - \Delta t)}{2\Delta t} + \mathcal{O}(\Delta t^2)$$